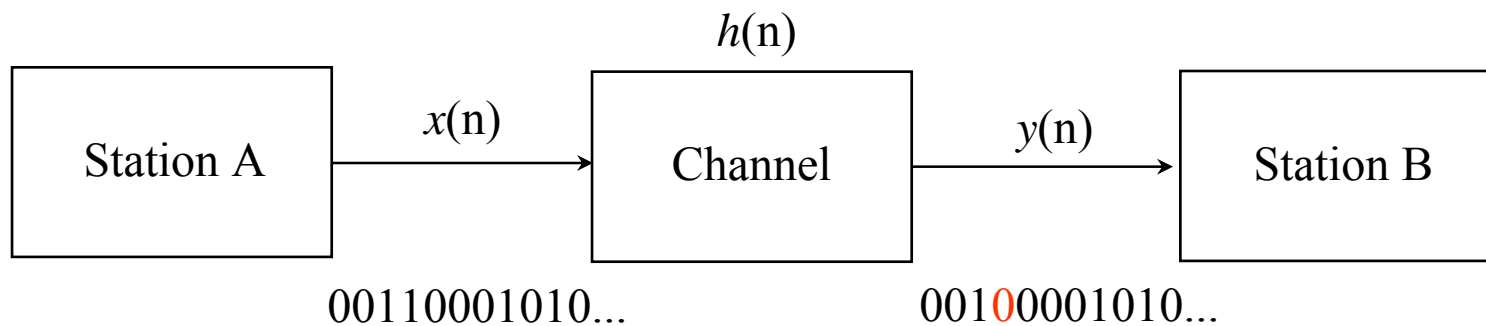


# Typical Electrical Engineering Problems

## A noisy binary communication channel

- The channel can be twisted pair, coaxial cable, fiber optic cable, or wireless medium.
- The channel introduces noise and thereby bit errors.



# Typical Electrical Engineering Problems

## Signal detection

- Desired target signal is buried in noise.

$$x(t) = A(t) \cos(\omega t + \phi(t)) + n(t)$$

- Determine the presence or absence of the desired signal.
- Filter the signal out of noise.
- Demodulate the signal.

# Typical Electrical Engineering Problems

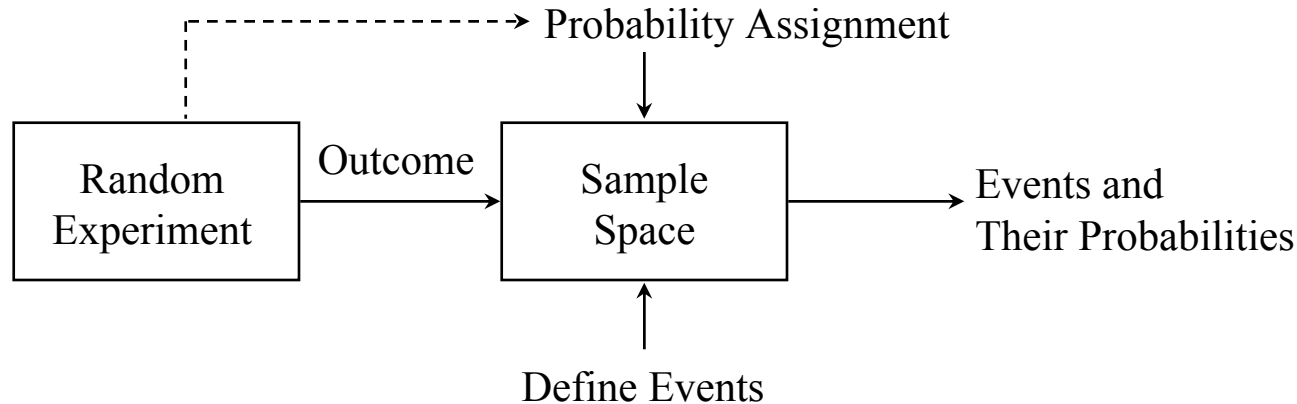
## Networks

- In large computer networks, there are limited resources (e.g., bandwidth, routers, switches, printers and other devices) that need to be shared by the users.
  - User jobs/packets are queued and assigned service based on predefined criteria.
  - Demand is uncertain and service time is also uncertain.
  - Delay from the time the service is requested to the time it is completed.
- Telephone networks, multiuser computer networks, and other communication networks.

# Our Interest and Goals

- Study tools to characterize the uncertainty
  - Probability theory, random variables, random processes
- Apply the tools to characterize non-deterministic signals
  - Random events, random signals
- Analyze systems processing non-deterministic signals
  - LTI systems with random inputs
  - Communication channels with noise
  - Communication networks with uncertain delays

# Basics of Probability



- In a random experiment, the outcome is uncertain
  - Physical experiment
  - Abstraction
- The entire collection of outcomes is the sample space,  $S$ 
  - Universal event or certain event
- An event consists of a single or a group of outcomes.
  - Events are user defined:  $A, B, \dots$
- A measure of likelihood of occurrence of an event,  $A$ 
  - Probability of  $A$  or  $\Pr[A]$

## **Example:**

Roll a die once. All faces are equally likely.

- Sample space [discrete sample space]

$$S = \{1,2,3,4,5,6\}$$

- Define events:

$$A_1 = \{\text{Odd numbered face}\} = \{1,3,5\}$$

$$A_2 = \{\text{Face value} < 3\} = \{1,2\}$$

$$A_3 = \{\text{Even numbered face value}\} = \{2,4,6\}$$

- Probability assignment:

$$\Pr[A_1] = 3/6 = 1/2$$

$$\Pr[A_2] = 2/6 = 1/3$$

$$\Pr[A_3] = 3/6 = 1/2$$

# Continuous Sample Space

Continuous sample space  $\Leftrightarrow$  Continuous magnitude signals



- Output of the radio receiver is measured at  $t = t_1$ . The dynamic range of the receiver output is  $-5V$  to  $+5V$   
$$S = \{s: -5 \leq s \leq +5\}$$
- A continuous sample space has uncountably infinite values or outcomes
  - $s$  could take values like 4.9326784531432677...
- Examples of events and probability assignments:

$$A_1 = \{s : -2.5 \leq s \leq 2.5\}, \quad \Pr[A_1] = 0.50$$

$$A_2 = \{s : -1 \leq s \leq 1\}, \quad \Pr[A_2] = 0.20$$

$$A_3 = \{s : s = 2.3 + \Delta x\}, \quad \lim_{\Delta x \rightarrow 0} \Pr[A_3] = 0$$

## Discrete Sample Space

Discrete sample space  $\Leftrightarrow$  Discrete valued signals



- Output of an 8-bit ADC contains only  $2^8 = 256$  values

$S = \{-5, -4.9609375, \dots, -0.0390625, 0, 0.0390625, \dots, 4.9609375\}$  or  
 $S = \{-128, -127, \dots, -1, 0, 1, \dots, 127\}$ : decimal equivalent of 2's complement representation

- A discrete sample space has finite or countably infinite values or outcomes
  - In this case, we have 256 values or outcomes (finite)



## Two Dimensional Sample Space

Roll two dice [discrete sample space]

$$S = \{(i,j): (1,1), (1,2), (1,3), \dots, (4,3), \dots, (6,4), (6,5), (6,6)\}$$

6	○	○	○	○	○	○
5	○	○	○	○	○	○
4	○	○	○	○	○	○
3	○	○	○	○	○	○
2	○	○	○	○	○	○
1	○	○	○	○	○	○
	1	2	3	4	5	6

# Events and Event Operations

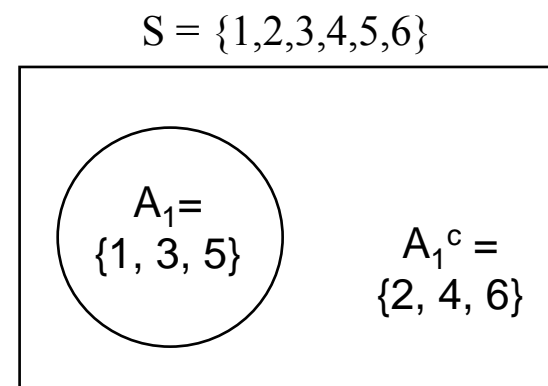
$S$	Certain event
$\emptyset$	Null event
$A_1, A_2, A_3, \dots$	User defined events
$A_1 + A_2$	Union operation ( $A_1 \cup A_2$ )
$A_1 A_2$	Intersection operation ( $A_1 \cap A_2$ )
$A^C$	Complement operation

## Example:

Consider a sample space represented by  $S = \{1,2,3,4,5,6\}$

Let  $A_1 = \{1,3,5\}$ ,  $A_2 = \{1,2\}$ , and  $A_3 = \{2,4,6\}$  be the user defined events

- $A_1^c = (\{1,3,5\})^c = \{2,4,6\} = A_3$
- $A_2 + A_3 = \{1,2,4,6\}$
- $A_1 + A_3 = \{1,2,3,4,5,6\} = S$
- $A_1 A_2 = \{1\}$ ,  $A_2 A_3 = \{2\}$
- $A_1 A_3 = 0$
- $A_1 + 0 = A_1$
- $A_1 0 = 0$



## Postulates for the Algebra of Events

- $A_1 A_1^C = 0$  Mutual exclusion
- $A_1 S = A_1$  Inclusion
- $(A_1^C)^C = A_1$  Double complement
- $A_1 + A_2 = A_2 + A_1$  Commutative law
- $A_1 + (A_2 + A_3) = (A_1 + A_2) + A_3$  Associative law
- $A_1 (A_2 + A_3) = A_1 A_2 + A_1 A_3$  Distributive law
- $(A_1 A_2)^C = A_1^C + A_2^C$  De Morgan's law

## Other Identities

$$- S^C = 0$$

$$- A_1 + 0 = A_1$$

$$- A_1 A_2 = A_2 A_1$$

$$- A_1 (A_2 A_3) = (A_1 A_2) A_3$$

$$- A_1 + (A_2 A_3) = (A_1 + A_2) (A_1 + A_3)$$

$$- (A_1 + A_2)^C = A_1^C A_2^C$$

Inclusion

Commutative law

Associative law

Distributive law

De Morgan's law

## Finite Unions and Intersections

These are included in the algebra.

$$\bigcup_{i=1}^N A_i = A_1 + A_2 + A_3 + \dots + A_N$$
$$\bigcap_{i=1}^N A_i = A_1 A_2 A_3 \dots A_N$$

## Infinite Unions and Intersections

If they are included, the algebra of events is called a Sigma Algebra.

$$\bigcup_{i=1}^{\infty} A_i = A_1 + A_2 + A_3 + \dots$$
$$\bigcap_{i=1}^{\infty} A_i = A_1 A_2 A_3 \dots$$

## Mutually Exclusive and Collectively Exhaustive Sets of Events

Mutually exclusive:  $A_k \cap A_j = \emptyset \quad k \neq j$

Collectively exhaustive:  $\bigcup_j A_j = S$

### Working Definition of the Sample Space

**THE SAMPLE SPACE IS REPRESENTED BY THE FINEST GRAIN, MUTUALLY EXCLUSIVE, COLLECTIVELY EXHAUSTIVE SET OF OUTCOMES FOR AN EXPERIMENT.**

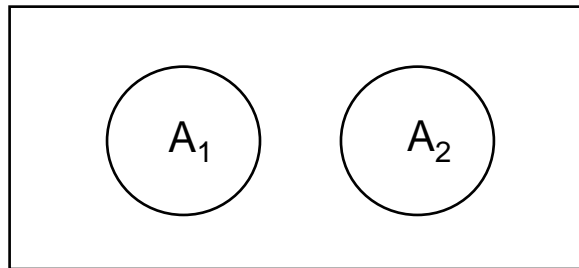
# The Axioms of Probability

I.  $\Pr[A_i] \geq 0$  for any event

II.  $\Pr[S] = 1$

III(a). If  $A_1 A_2 = 0$ , then  $\Pr[A_1 + A_2] = \Pr[A_1] + \Pr[A_2]$

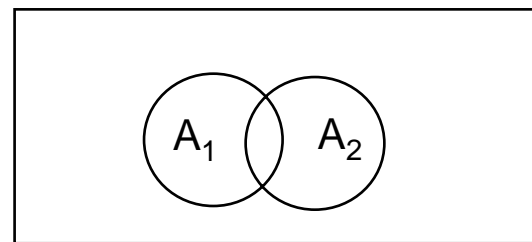
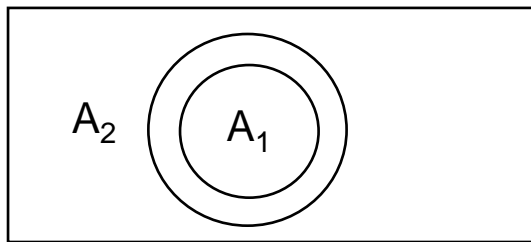
III(b). If  $A_i A_j = 0$  for  $i \neq j$ , then  $\Pr\left[\bigcup_{i=1}^{\infty} A_i\right] = \sum_{i=1}^{\infty} \Pr[A_i]$





## Some Corollaries

1.  $\Pr[A^c] = 1 - \Pr[A]$  (ii, iii)
2.  $0 \leq \Pr[A] \leq 1$  (i, ii, iii)
3. If  $A_1 \subset A_2$ , then  $\Pr[A_1] \leq \Pr[A_2]$  (i, iii)
4.  $\Pr[0] = 0$  (ii, iii)
5. If  $A_1 A_2 = 0$ , then  $\Pr[A_1 A_2] = 0$
6.  $\Pr[A_1 + A_2] = \Pr[A_1] + \Pr[A_2] - \Pr[A_1 A_2]$



# The Principle of Total Probability

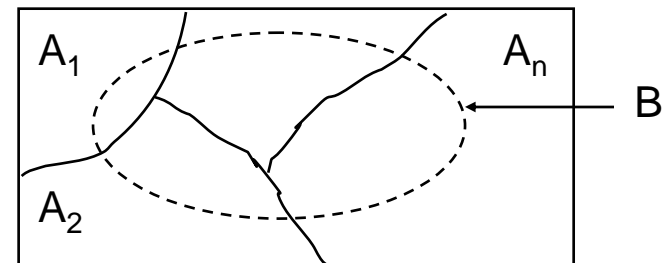
Let  $A_1, A_2, \dots, A_n$  be a set of mutually exclusive and collectively exhaustive events:

$$A_k \cap A_j = \emptyset \quad k \neq j$$

$$\bigcup_{j=1}^n A_j = S \quad \text{then} \quad \sum_{j=1}^n \Pr[A_j] = 1$$

Now let  $B$  be any event in  $S$ . Then,

$$\Pr[B] = \Pr[BA_1] + \Pr[BA_2] + \dots + \Pr[BA_n]$$



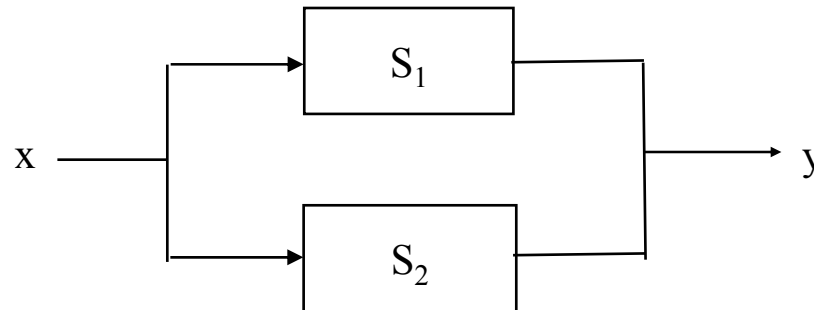
# Independence of Events

Two events  $A_1$  and  $A_2$  are said to be statistically independent if and only if

$$\Pr[A_1 A_2] = \Pr[A_1] \Pr[A_2]$$

# System Reliability Calculations

## Parallel Connection of switches



Define:  $A_1 = \{S_1 \text{ fails}\}$ ,  $\Pr[A_1] = p$ ,  $\Pr[A_1^c] = 1 - p = q$

$A_2 = \{S_2 \text{ fails}\}$ ,  $\Pr[A_2] = p$ ,  $\Pr[A_2^c] = 1 - p = q$

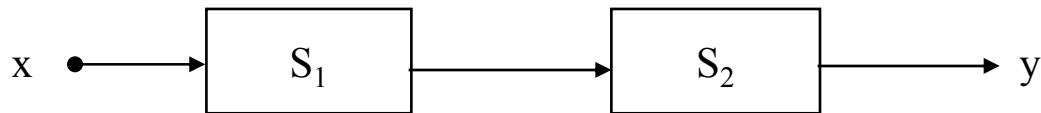
$F = \{\text{no connection between } x \text{ and } y\} = A_1 A_2$

The probability that the connection fails:

$$\begin{aligned}\Pr[F] &= \Pr[A_1 A_2] = \Pr[A_1] \Pr[A_2] & A_1 \text{ and } A_2 \text{ are independent} \\ &= p^2\end{aligned}$$

# System Reliability Calculations

## Series connection of switches



Assume that switch failures are statistically independent; failure results in an open connection.

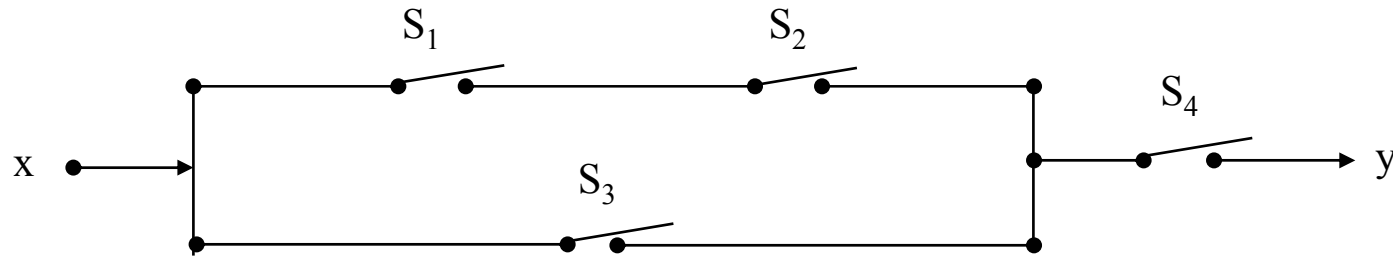
$$F = \{\text{no connection between } x \text{ and } y\} = A_1 + A_2$$

The probability that the connection fails:

$$\begin{aligned} \Pr[F] &= \Pr[A_1 + A_2] = \Pr[A_1] + \Pr[A_2] - \Pr[A_1 A_2] \\ &= \Pr[A_1] + \Pr[A_2] - \Pr[A_1] \Pr[A_2] \quad A_1 \text{ and } A_2 \text{ are independent} \\ &= p + p - p^2 = 2p - p^2 \end{aligned}$$

## Example:

Consider a simple switching network as follows:



Define:  $A_k = \{\text{switch } S_k \text{ fails}\}$ ,  $k = 1, 2, 3, 4$        $\Pr[A_k] = p$

(a) Find the probability that the path between x and y is established.

Let  $F = \{\text{no connection between x and y}\} = (A_1 + A_2)A_3 + A_4$

The probability of path failure is given by

$$\begin{aligned}\Pr[F] &= \Pr[(A_1 + A_2)A_3 + A_4] = \Pr[A_1A_3 + A_2A_3] + \Pr[A_4] - \Pr[A_1A_3A_4 + A_2A_3A_4] \\ &= \Pr[A_4] + \Pr[A_1A_3] + \Pr[A_2A_3] - \Pr[A_1A_2A_3] \\ &\quad - \Pr[A_1A_3A_4] - \Pr[A_2A_3A_4] + \Pr[A_1A_2A_3A_4] \\ &= p + 2p^2 - 3p^3 + p^4\end{aligned}$$

The desired probability is then given by

$$\Pr[\text{path established}] = 1 - \Pr[F] = 1 - p - 2p^2 + 3p^3 - p^4$$

(b) Compute the desired probability as a function of p:

$$\Pr[\text{path established}] = 1 - \Pr[F] = 1 - p - 2p^2 + 3p^3 - p^4$$

p	1 - Pr[F]
0.1	0.8829
0.01	0.98980299
0.001	0.998998002999
0.0001	0.999899980003

## Repeated Independent Trials (Bernoulli Trials)

A random experiment,  $E$ , consists of several sub-experiments or trials,  $E_i$ :

- All sub-experiments have the same sample space,  $S_i$ .
- Events from all sub-experiments are mutually independent:

$$\Pr[A_1 A_2 A_3 \dots A_n] = \Pr[A_1] \Pr[A_2] \Pr[A_3] \dots \Pr[A_n],$$

where  $A_i$  is an event from  $S_i$ .

- The sample space of  $E$  is:

$$S = S_1 \times S_2 \times S_3 \times \dots \times S_n$$



## Counting Methods and Probability

The assignment of probability is given by

$$\Pr[A_1] = \frac{\text{Number of outcomes in Event } A_1}{\text{Number of outcomes of experiment}}$$

### The rule of products:

Consider an experiment with  $n$  outcomes; repeat  $r$  times

Total number of outcomes is given by

$$N_r^n = n \cdot n \cdots n = n^r$$

or in a general case

$$N_r^{n_i} = n_1 n_2 \cdots n_r = \prod_{i=1}^r n_i$$

**Example:**

Roll a die 4 times sequentially. The total number of outcomes is

$$N_4^6 = 6 \cdot 6 \cdot 6 \cdot 6 = 6^4 = 1296$$

**Example:**

Form 5-letter words using 26 English alphabet characters. Characters can be repeated, and the words so formed do not have to be meaningful.

$$N_5^{26} = 26^5 = 11,881,376$$

**Example:**

Construct variable names of length 3 using a letter, a number, and a letter (e.g., A2C).

$$N_3^{n_i} = 26 \cdot 10 \cdot 26$$

## PERMUTATIONS (products without replacement)

Select  $r$  objects from among a given set of  $n$  distinct objects where we pay attention to the order in which the  $r$  objects are selected.

$$\begin{aligned} P_r^n &= n(n-1)(n-2)\cdots(n-r+1) \\ &= \frac{n!}{(n-r)!}, \quad \text{for } r \leq n \end{aligned}$$

Special case: For  $r = n$ :  $P_n^n = n!$

## **Example:**

Form 5-letter words using the English alphabet. The characters cannot be repeated, and the words do not have to be meaningful.

The total number of words that can be formed is:

$$\begin{aligned} P_5^{26} &= \frac{26!}{(26-5)!} \\ &= \frac{26!}{21!} = 26 \cdot 25 \cdot 24 \cdot 23 \cdot 22 = 7,893,600 \end{aligned}$$

## COMBINATIONS (without replacement, without order)

Select  $r$  objects from among a given set of  $n$  distinct objects where we pay no attention to the order in which the  $r$  objects are selected.

$$C_r^n = \frac{n(n-1)\cdots(n-r+1)}{r(r-1)(r-2)\cdots(1)} = \frac{n!}{r!(n-r)!} = \binom{n}{r}$$

Binomial coefficient      “ $n$  choose  $r$ ”

## **Example:**

Consider 5 workstations having equal capabilities:  $\{a, b, c, d, e\}$

- *Permutations*: Select two workstations where one will be a server and the other a graphics workstation. The possible selections are:

ab ba ac ca ad da ae ea bc cb bd db be eb cd dc ce  
ec de ed

$$P_2^5 = \frac{5!}{(5-2)!} = 5 \cdot 4 = 20 \quad \text{“ab} \neq \text{ba”}$$

- *Combinations*: Select two workstations where both will be used as graphics workstations. The possible selections are:

ab ac ad ae bc bd be cd ce de

$$C_2^5 = \frac{5!}{2!(5-2)!} = \frac{5 \cdot 4}{2} = 10 \quad \text{“ab} = \text{ba”}$$

## **Example:**

We plan to buy 5 personal computers. The computer store has a stock of 10 foreign made PCs and 15 US made PCs that meet our specifications.

(a) Assuming that the 5 computers are randomly chosen from this lot, what is the probability that exactly 3 US made computers are selected?

- The sample space is given by

$$S = \{\text{combinations of } r = 5 \text{ chosen from } n = 25\}$$

$$N_S = C_5^{25} = \frac{25!}{5!(25-5)!}$$

- The desired event  $A = \{\text{exactly 3 of the 5 selected are US made}\}$

$$N_A = C_3^{15} C_2^{10} = \frac{15! 10!}{3!(15-3)! 2!(10-2)!}$$

- The probability that we have 3 US made computers is  
[Hyper geometric distribution]

$$\Pr[A] = \frac{N_A}{N_S} = \frac{C_3^{15} C_2^{10}}{C_5^{25}} = \frac{15! 10! 5! (25-5)!}{3! (15-3)! 2! (10-2)! 25!} \cong 0.3854$$

### **Example (continued):**

(b) Assuming that the 5 computers are randomly chosen from this lot, what is the probability that *at least* 1 is foreign made?

- Let event  $B = \{\text{none of the 5 selected computers is foreign made}\}$
- The desired event  $C = \{\text{one or more of the 5 are foreign made}\}$
- Since  $B^C = C$ , we can write  $\Pr[C] = 1 - \Pr[B]$

$$\Pr[B] = \frac{N_B}{N_S} = \frac{C_5^{15} C_0^{10}}{C_5^{25}} = \frac{15!}{5!} \frac{10!}{(15-5)!} \frac{5!}{0!} \frac{(25-5)!}{(10-0)!} \frac{1}{25!} \cong 0.05652$$

$$\Pr[C] = 1 - \Pr[B] = 0.94348$$



## **Example:**

Consider a box of 25 modem chips: 5 of them are known to be defective. Select 6 from the box at random and test them. What is the probability that exactly 2 are defective?

- Sample Space:  $S = \{\text{combinations of } r = 6 \text{ chosen from } n = 25\}$
- Event:  $A = \{\text{exactly 2 of the 6 selected chips are defective}\}$
- The number of outcomes in  $S$  is given by:

$$N_S = C_6^{25} = \frac{25!}{6!(25-6)!}$$

- For the selected 6 chips, we are interested in the case, where 2 are defective (i.e., they are from the 5 defective chips in the box) and 4 are non-defective (i.e., they are from the 20 non-defective chips in the box).

### **Example (continued):**

- The number of outcomes in A is given by:

$$N_A = C_2^5 C_4^{20} = \frac{5! 20!}{2!(5-2)! 4!(20-4)!}$$

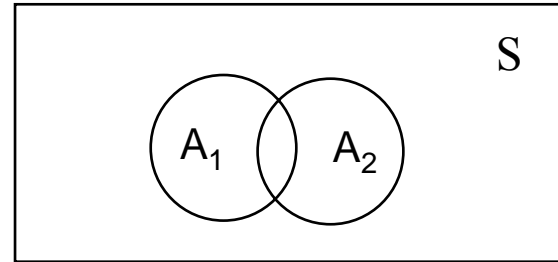
- The probability that exactly 2 of the 6 selected chips are defective is [Hyper geometric distribution]

$$\begin{aligned} \Pr[A] &= \frac{N_A}{N_S} = \frac{C_2^5 C_4^{20}}{C_6^{25}} \\ &= \frac{5! 20! 6! (25-6)!}{2! (5-2)! 4! (20-4)! 25!} \cong 0.2736. \end{aligned}$$

# Conditional Probability

Probability of occurrence of one event (say,  $A_1$ ) subject to the knowledge that another event (say,  $A_2$ ) has occurred.

$$\Pr[A_1 | A_2] = \frac{\Pr[A_1 \cap A_2]}{\Pr[A_2]}$$



$\Pr[A_1 | A_2]$  is read as “probability of  $A_1$  *given*  $A_2$ ”

If  $A_1$  and  $A_2$  are independent, then

$$\Pr[A_1 | A_2] = \frac{\Pr[A_1 \cap A_2]}{\Pr[A_2]} = \frac{\Pr[A_1] \Pr[A_2]}{\Pr[A_2]} = \Pr[A_1]$$

## Example:

Consider a sequence of 3 binary numbers (occurring randomly).

Sample space:  $S = \{000, 001, 010, 011, 100, 101, 110, 111\}$

What is the probability that there are more 1's than 0's given that the first bit is a 1.

- Let us define two events:

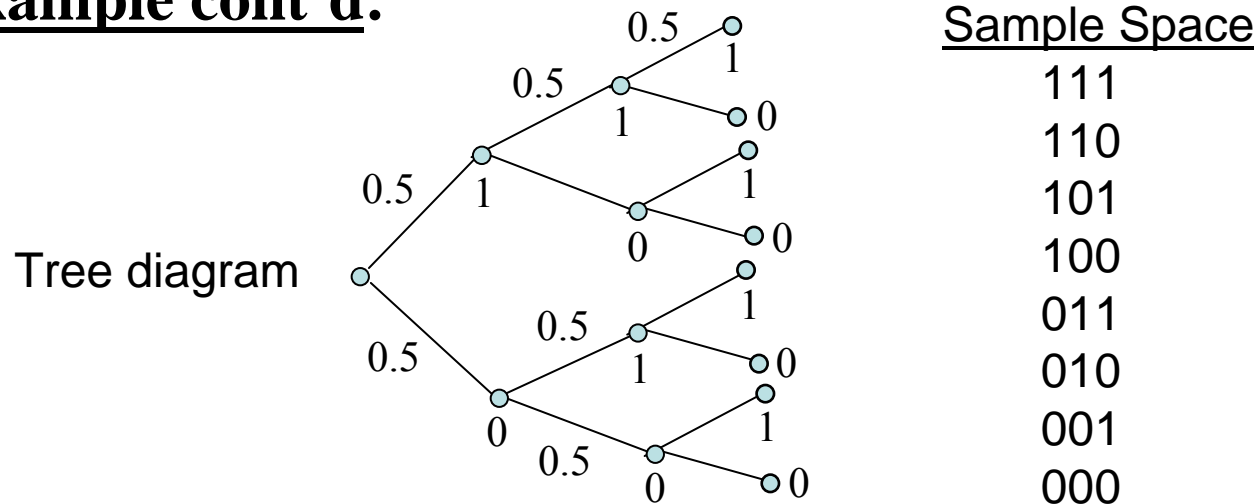
$$A_1 = \{\text{more 1's than 0's}\} = \{011, 101, 110, 111\}$$

$$A_2 = \{\text{the first bit is a 1}\} = \{100, 101, 110, 111\}$$

- Their intersection:

$$A_1 A_2 = \{101, 110, 111\}$$

### Example cont'd:



- All 8 events in the sample space have probability  $1/8$ , therefore

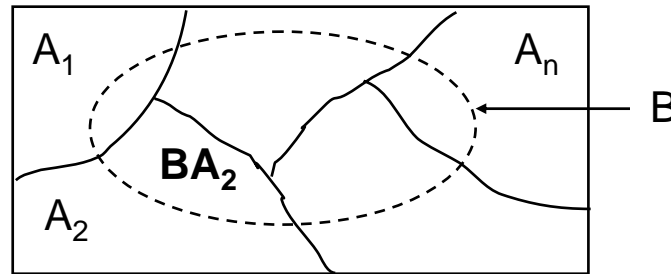
$$\Pr[A_2] = \frac{4}{8} \quad \text{and} \quad \Pr[A_1 A_2] = \frac{3}{8}$$

- The conditional probability is obtained as follows:

$$\Pr[A_1 | A_2] = \frac{\Pr[A_1 A_2]}{\Pr[A_2]} = \frac{3/8}{4/8} = \frac{3}{4}$$

# The Principle of Total Probability Revisited

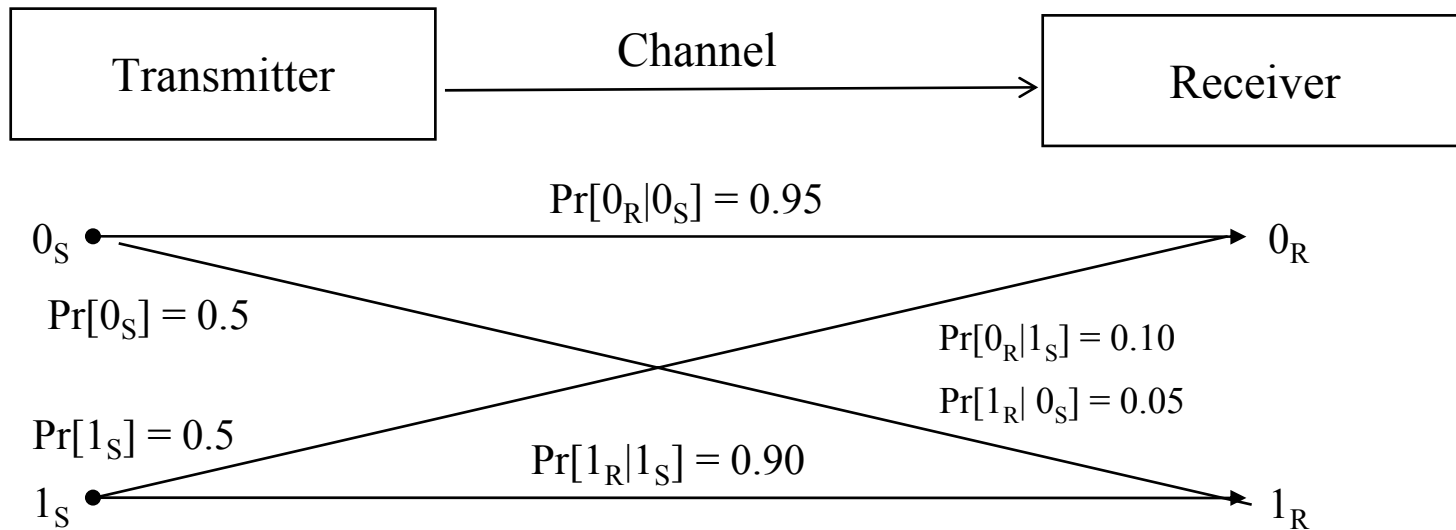
Let  $A_1, A_2, \dots, A_n$  be mutually exclusive and collectively exhaustive events. Let  $B$  be an event in  $S$ .



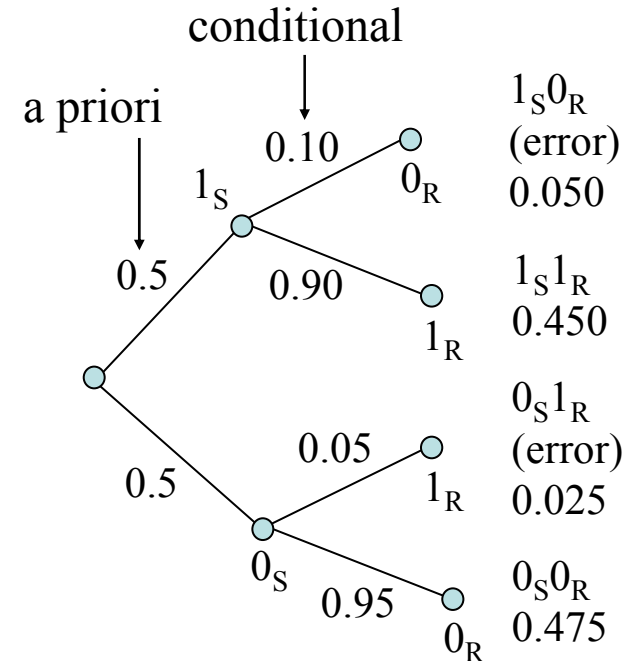
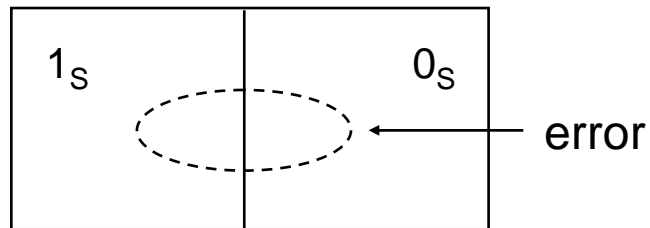
Then,

$$\begin{aligned}\Pr[B] &= \Pr[BA_1] + \Pr[BA_2] + \dots + \Pr[BA_n] \\ &= \Pr[B \mid A_1] \Pr[A_1] + \dots + \Pr[B \mid A_n] \Pr[A_n] \\ &= \sum_{i=1}^n \Pr[B \mid A_i] \Pr[A_i]\end{aligned}$$

## Example: Binary Communication Channel



## Example cont'd.



$$\Pr[\text{error} | 1_S] = \Pr[0_R | 1_S] = 0.10$$

$$\Pr[\text{error} | 0_S] = \Pr[1_R | 0_S] = 0.05$$

$$\begin{aligned}\Pr[\text{error}] &= \Pr[\text{error} | 1_S] \Pr[1_S] + \Pr[\text{error} | 0_S] \Pr[0_S] \\ &= 0.10 \cdot 0.50 + 0.05 \cdot 0.50 = 0.075\end{aligned}$$



## More on Conditional Probability

From the definition of conditional probability, we can write

$$\Pr[A_1|A_2] = \frac{\Pr[A_1 \cap A_2]}{\Pr[A_2]} \quad \text{or} \quad \Pr[A_1 \cap A_2] = \Pr[A_1|A_2] \Pr[A_2]$$

$$\Pr[A_2|A_1] = \frac{\Pr[A_1 \cap A_2]}{\Pr[A_1]} \quad \text{or} \quad \Pr[A_1 \cap A_2] = \Pr[A_2|A_1] \Pr[A_1]$$

$$\therefore \Pr[A_1|A_2] \Pr[A_2] = \Pr[A_2|A_1] \Pr[A_1]$$

This can be written as

$$\Pr[A_2|A_1] = \frac{\Pr[A_1|A_2] \Pr[A_2]}{\Pr[A_1]}$$

## Bayes' Rule

Let  $A_1, A_2, \dots, A_n$  be a set of mutually exclusive, collective exhaustive events. Then,

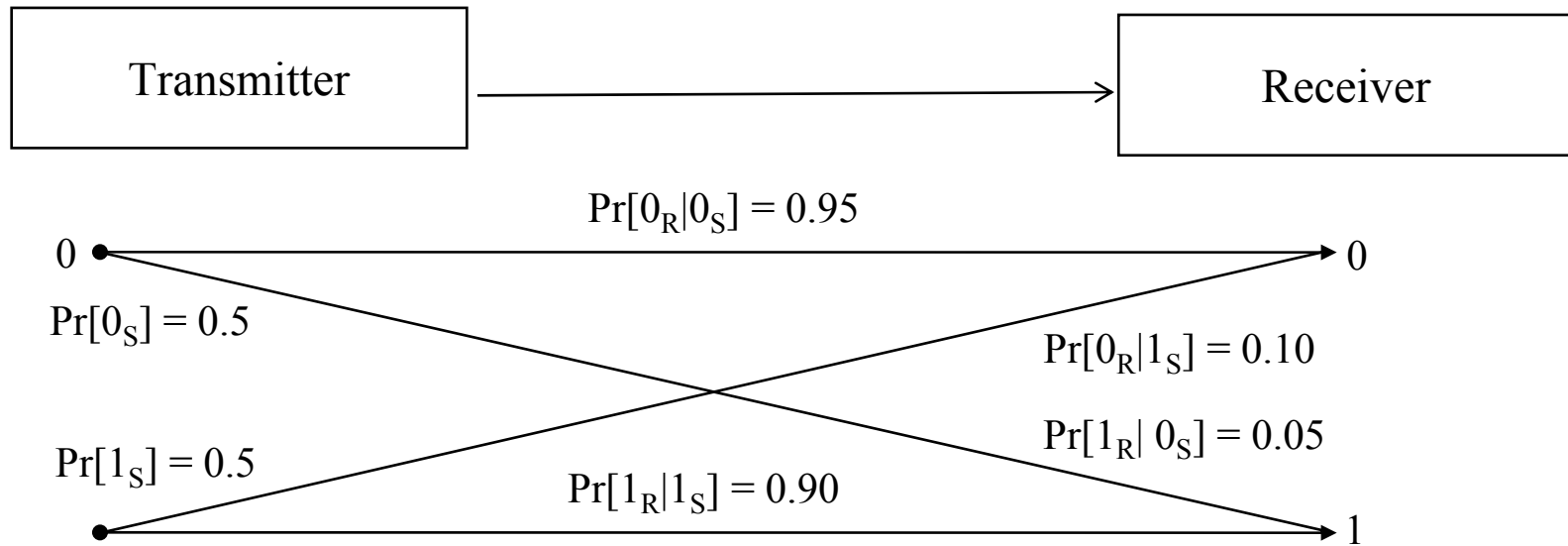
$$\Pr[A_j | B] = \frac{\Pr[B | A_j] \Pr[A_j]}{\Pr[B]}$$

Or, applying the Principle of Total Probability

$$\Pr[A_j | B] = \frac{\Pr[B | A_j] \Pr[A_j]}{\sum_{k=1}^n \Pr[B | A_k] \Pr[A_k]}$$

This is called Bayes' Rule.

## Example: Binary Communication Channel



Determine the inverse probability,  $P[1_S | 1_R]$ .

## Example cont'd.

Determine the inverse probability,  $P[1_S | 1_R]$ :

$$\begin{aligned}\Pr[1_S | 1_R] &= \frac{\Pr[1_R | 1_S] \Pr[1_S]}{\Pr[1_R]} \\ &= \frac{\Pr[1_R | 1_S] \Pr[1_S]}{\Pr[1_R | 1_S] \Pr[1_S] + \Pr[1_R | 0_S] \Pr[0_S]} \\ &= \frac{0.45}{0.45 + 0.025} = 0.9474\end{aligned}$$

## **Example:**

Consider 3 boxes of ICs:

Box 1 contains 1500 ICs and 10% of them are defective;

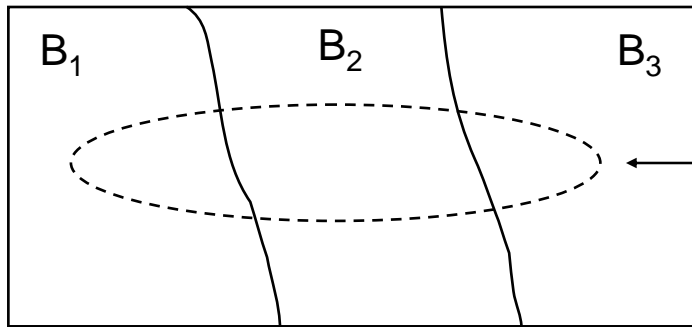
Box 2 contains 2000 ICs and 20% of them are defective; and

Box 3 contains 3000 ICs and 16% of them are defective.

Select 1 of the 3 boxes at random and choose an IC from that box at random.

(a) What is the probability that this IC is defective?

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Define:  $A$  = “selected IC is defective” ,  
 $B_i$  = “IC is from box  $i$ ”

- By the principle of total probability, we can write

$$\begin{aligned}\Pr[A] &= \Pr[A|B_1]\Pr[B_1] + \Pr[A|B_2]\Pr[B_2] + \Pr[A|B_3]\Pr[B_3] \\ &= 0.10 \cdot \frac{1}{3} + 0.20 \cdot \frac{1}{3} + 0.16 \cdot \frac{1}{3} = \frac{0.46}{3} = 0.1533\end{aligned}$$

- (b) Suppose that the selected IC is found to be defective.  
What is the probability that this IC came from box #3?

By Bayes' theorem, we can write

$$\begin{aligned}\Pr[B_3 | A] &= \frac{\Pr[A | B_3] \Pr[B_3]}{\Pr[A]} \\ &= \frac{0.16 \cdot \frac{1}{3}}{\frac{0.46}{3}} = \frac{0.16}{0.46} = 0.3478\end{aligned}$$

- (c) Suppose all IC's are thoroughly mixed in one box and an IC is selected at random from the box. What is the probability that the IC is defective?

## Basic Information Theory

Given an event  $A$  and its probability  $\Pr[A]$ , information associated with  $A$  is given by

$$I[A] = \log_x \frac{1}{\Pr[A]} = -\log_x \Pr[A],$$

where  $x$  is the base of the logarithm:

if  $x = 2$ , the units of information are bits;

$x = 10$ , the units are hartleys;

and  $x = e$ , the units are nats.

Note the identity:  $\log_a b = x$  means that  $a^x = b$ .



## **Example:**

Consider two events:  $A_1$  and  $A_2$  with corresponding probabilities of occurrence of 0.125 and 0.875, respectively.

The information associated with these events:

$$I[A_1] = -\log_2 (0.125) = 3 \text{ bits and } I[A_2] = -\log_2 (0.875) = 0.1925 \text{ bits.}$$

## **Entropy**

Given a set of independent events that are mutually exclusive and collectively exhaustive, we can define the average information associated with the random experiment as

$$H = \sum_i \Pr[A_i] \cdot I[A_i] = - \sum_i \Pr[A_i] \cdot \log_x \Pr[A_i].$$

## **Example:**

Consider a sequence 1 2 3 2 3 4 5 4 5 6 7 8 9 8 9 0.

We estimate the probability of occurrence of each symbol as follows:

$$\Pr[1] = \Pr[6] = \Pr[7] = \Pr[0] = 1/16$$

$$\Pr[2] = \Pr[3] = \Pr[4] = \Pr[5] = \Pr[8] = \Pr[9] = 2/16.$$

The entropy of this sequence is

$$H = -\sum_i \Pr[A_i] \cdot \log_2 \Pr[A_i] = 3.25 \text{ bits.}$$

# Shannon-Fano Code

- Messages are composed of an alphabet in which the frequency of occurrence of each letter is a probabilistic phenomenon.
  - For transmission purposes the messages are compressed such that the code length of a letter is inversely proportional to its frequency of occurrence (e.g., think of the Morse code).
  - Since the letters are transmitted sequentially, no short codeword be part of the start of a longer codeword for unique decodability.
- Shannon-Fano Algorithm:
  - Arrange letters in a descending order of their probabilities by breaking any ties arbitrarily.
  - Starting at the top, partition the letters into two equi-probable subgroups (as closely as possible): assign 0 to the first subgroup and 1 to the second.
  - Continue partitioning the subgroups until all letters are exhausted: after each partition, assign a 0 to the first group and a 1 to the second and append the newly assigned bits to the previously assigned bits.

## Example

Given a text message “ELECTRICAL ENGINEERING,” determine the relative probabilities of the letters in the message and find the Shannon-Fano code for each letter. Ignore the space character.

- Since there are 21 letters in the message, we have the following probabilities:

Letters: {E, L, C, T, R, I, A, N, G}

Probabilities:  $\{5/21, 2/21, 2/21, 1/21, 2/21, 3/21, 1/21, 3/21, 2/21\}$

- Code assignment:

Code assignment:					Codeword	Length
E	5/21	0	0	(2)	00	2
I	3/21	0	1	(3)	010	3
N	3/21	0	1	(1)	011	3
L	2/21	1	0	(3)	100	3
C	2/21	1	0	(4)	1010	4
R	2/21	1	0	(2)	1011	4
G	2/21	1	1	(3)	110	3
T	1/21	1	1	(4)	1110	4
A	1/21	1	1		1111	4

## The Binomial Probability Law

Consider a sequence of  $n$  binary values:

$$\Pr[1] = p, \Pr[0] = 1 - p = q.$$

Define:  $A = \{\text{occurrence of } r \text{ 1's in a sequence of length } n\}$

The number of ways  $r$  1's can occur in a sequence of length  $n$  is given by the binomial coefficient,  $C_r^n$ :

- Note that each of these arrangements has  $r$  1's and  $n - r$  0's.
- The probability of occurrence of such an arrangement is then given by  $p^r q^{n-r}$ .

The probability of the desired event:  $\Pr[A] = \binom{n}{r} p^r q^{n-r}$

### **Example 1:**

Consider a modem connection with a channel bit error rate  $p = 10^{-2}$ . Given that the data are sent as packets of 100 bits, what is the probability that (a) 1 bit is in error and (b) 3 bits are in error?

$$(a) \Pr[1 \text{ bit in error}] = \binom{100}{1} 0.01^1 \cdot 0.99^{99} = 0.3697$$

$$(b) \Pr[3 \text{ bits in error}] = \binom{100}{3} 0.01^3 \cdot 0.99^{97} = 0.060999$$

## **Example 2:**

Consider a communication system with a channel bit error rate,  $p = 10^{-3}$ . The transmitter sends each bit three times, and the receiver takes a majority poll of the received bits to determine the received bit. What is the probability of bit error now?

(a) Each transmission is a Bernoulli trial with  $n = 3$ .

Define:  $A = \{2 \text{ or more bit errors in } 3 \text{ trials}\}$

$$\begin{aligned}\Pr[\text{error}] &= \Pr[A] = \Pr[r \geq 2] \\ &= \binom{3}{2} p^2 (1 - p) + \binom{3}{3} p^3 \\ &= 3 \times 10^{-6} (1 - 10^{-3}) + 10^{-9} \cong 3 \times 10^{-6}\end{aligned}$$

## **Example 2 cont'd:**

(b) Consider 5 bits ( $n = 5$ ).

Define:  $A = \{3 \text{ or more bit errors in } 5 \text{ trials}\}$

$$\Pr[\text{error}] = \Pr[A] = \Pr[r \geq 3]$$

$$= \binom{5}{3} p^3 (1-p)^2 + \binom{5}{4} p^4 (1-p) + \binom{5}{5} p^5$$

$$\cong 9.985 \times 10^{-9}$$



### **Example:**

Consider a sequence of 10 binary digits. Let  $\Pr[1] = 0.52$ .

(a) What is the probability of obtaining 8 or more 1's?

Define:  $A = \{8 \text{ or more } 1\text{'s in a sequence of 10 bits}\}$

$$\begin{aligned}\Pr[A] &= \binom{10}{8} \cdot 0.52^8 \cdot 0.48^2 + \binom{10}{9} \cdot 0.52^9 \cdot 0.48 + \binom{10}{10} \cdot 0.52^{10} \\ &= 45 \cdot 0.52^8 \cdot 0.48^2 + 10 \cdot 0.52^9 \cdot 0.48 + 0.52^{10} \cong 0.0702161458426\end{aligned}$$

(b) What is the probability of obtaining exactly six 1's?

Define:  $A = \{\text{exactly six } 1\text{'s in a sequence of 10 bits}\}$

$$\Pr[A] = \binom{10}{6} \cdot 0.52^6 \cdot 0.48^4 = 210 \cdot 0.52^6 \cdot 0.48^4 \cong 0.220396303407$$

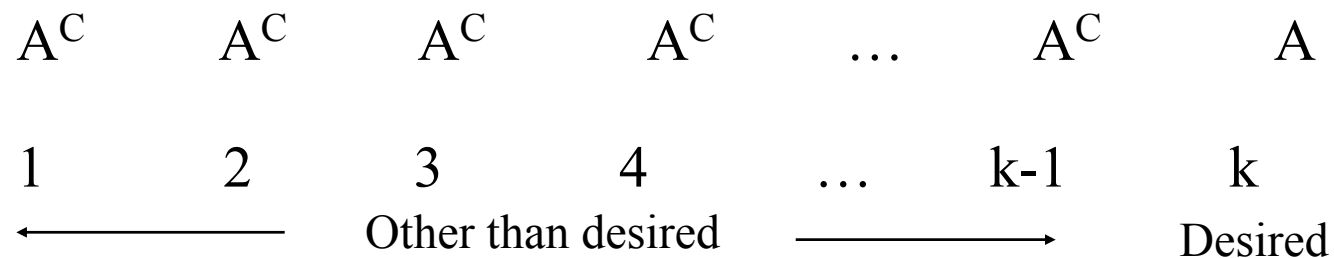
# The Geometric Probability Law

Consider a sub-experiment. Let  $A$  be the desired event.

Let  $\Pr[A] = p$ ,  $\Pr[A^C] = 1 - p$ .

Repeat the sub-experiment until  $A$  occurs.

- Suppose that  $A$  occurs in the  $k^{\text{th}}$  trial:



The probability that  $A$  occurs in the  $k^{\text{th}}$  trial is:

$$\begin{aligned}
 \Pr[A \text{ occurs in } k^{\text{th}} \text{ trial}] &= \underbrace{(1-p)(1-p)(1-p)(1-p)\cdots(1-p)}_{k-1 \text{ uneventful trials}} \quad p \\
 &= (1-p)^{k-1} \quad p
 \end{aligned}$$

## **Example:**

In a computer to computer modem link, the receiving computer has an error detection algorithm. If it detects a bit error, it requests retransmission of the packet. For simplicity, assume that the packet length is 8 bits. Let the probability of channel error be  $\Pr[\text{error}] = 0.1$ .

- (a) Determine the probability that the error occurs after the 5<sup>th</sup> bit in a packet.

$$\begin{aligned}\Pr[k > 5] &= \Pr[k = 6] + \Pr[k = 7] + \Pr[k = 8] \\ &= 0.9^5 \cdot 0.1 + 0.9^6 \cdot 0.1 + 0.9^7 \cdot 0.1 \\ &= 0.16002279\end{aligned}$$

## Example cont'd:

(b) What is the probability that a packet is retransmitted 2 times?

- A packet is retransmitted once if at least one of the 8 bits is in error.
- It is retransmitted again if at least one of the retransmitted 8 bits is in error.

$$\Pr[1 \text{ retransmission}] = \Pr[k \geq 1] = \sum_{i=1}^8 \Pr[k = i] = 0.5695$$

$$\Pr[2 \text{ retransmissions}] = (\Pr[1 \text{ retransmission}])^2 = 0.3244$$